

FACTOR 3

THE DIMENSIONALITY OF NATIONS PROJECT

RESEARCH REPORT

FACTOR 2

FACTOR 1

DEPARTMENT OF POLITICAL SCIENCE

UNIVERSITY OF HAWAII

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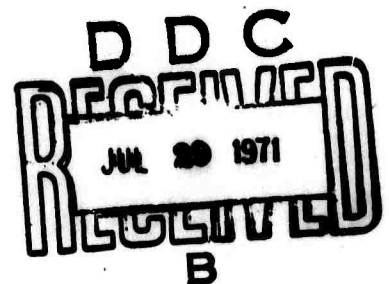
RESEARCH REPORT NO. 14
Computer Program Profile

Dennis R. Hall

JUN 1968

Prepared in Connection with the
National Science Foundation
Grant No. GS-1230
and the
Advanced Research Projects Agency
ARPA Order No. 1063
and Monitored by the
Office of Naval Research
Contract #N00014-67-A-0387-0003

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(Security classification of title, body of abstract and indexing annotation must be entered when the overall report is classified)

1. ORIGINATING ACTIVITY (Corporate author) The Dimensionality of Nations Project University of Hawaii 2500 Campus Road, Honolulu, Hawaii 96822		2a. REPORT SECURITY CLASSIFICATION Unclassified	
3. REPORT TITLE Computer Program Profile		2b. GROUP	
4. DESCRIPTIVE NOTES (Type of report and inclusive dates) Research Report No. 14			
5. AUTHOR(S) (First name, middle initial, last name) Dennis R. Hall			
6. REPORT DATE June 1968		7a. TOTAL NO. OF PAGES 24	7b. NO. OF REFS 0
8a. CONTRACT OR GRANT NO. N00014-67-A-0387-0003		9a. ORIGINATOR'S REPORT NUMBER(S) Research Report No. 14	
b. PROJECT NO.		9b. OTHER REPORT NO(S) (Any other numbers that may be assigned this report)	
c.			
d.			
10. DISTRIBUTION STATEMENT This document has been approved for public release and sale; its distribution is unlimited and reproduction in whole or in part is permitted for any purpose of the United States government.			
11. SUPPLEMENTARY NOTES 2500 Campus Road Honolulu, Hawaii 96822		12. SPONSORING MILITARY ACTIVITY Advanced Research Projects Agency Washington, D.C.	
13. ABSTRACT Computer techniques designed to construct numerical taxonomies from large, complex arrays of data have developed rapidly in the past few years. It has been the experience of many who have attempted to employ these techniques, however, that mathematically logical taxonomies were drawn. This difficulty has all too often left the researcher with a neat list of the sets of cases which are numerically similar, but with only the vaguest grasp of what the grouping similarities might be. This report introduces a simple computer technique to plot the underlying similarity of groups.			

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TABLE OF CONTENTS

I.	Introduction	2
II.	Development of the Grouping Problem . . .	2
III.	Application	5
IV.	General Description	8
V.	Profile Program Writeup	10
VI.	Fortran IV Source Listing	12
VII.	Sample Problem and Output	13

FIGURES

1.	Attribute Profile for France in 1955 . . .	3
2.	Dendrogram Mapping of Groups	7
3.	Attribute Profile for Group from Dendrogram Fig. 2	9

I INTRODUCTION

Computer techniques designed to construct numerical taxonomies from large, complex arrays of data have developed rapidly in the past few years. It has been the experience of many who have attempted to employ these techniques, however, that mathematically logical taxonomies are difficult to relate to the original data from which the taxonomies were drawn. This difficulty has all too often left the researcher with a neat list of the sets of cases which are numerically similar, but with only the vaguest grasp of what the grouping similarities might be. This report introduces a simple computer technique to plot the underlying similarity of groups.

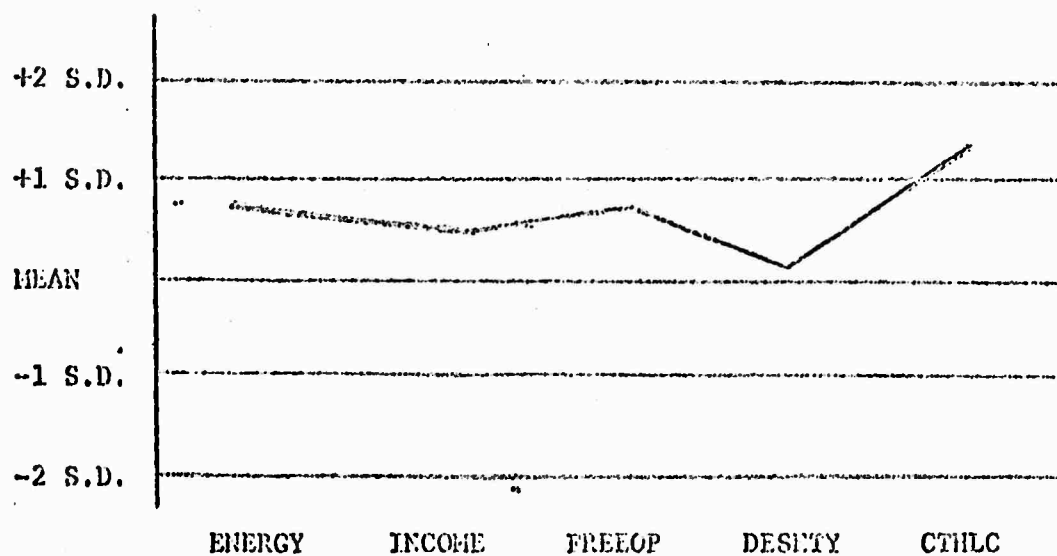
There are two basic types of numerical taxonomy; hierarchical, and cross-sectional. The former taxonomy -- the type generally employed in biology -- classifies life from a single instance of a living organism, through the species, phylum, kingdom, to all of organic existence. The hierarchical taxonomic scheme assumes dynamic similarity for grouping. A cross-sectional taxonomy, the type most generally employed by social scientists, assumes static similarity. A case is grouped as a member of one of k groups, each case is a member of one and only one group, and the group is assumed to be qualitatively distinct from all other groups.

II. DEVELOPMENT OF THE GROUPING PROBLEM

A Case is characterized by the values it has for a range of variables. These values may be specific, as the national income of France in U. S. dollars, or they may be general, as the freedom of group opposition in France measured by a two point scale. The numerical scores for these characteristics of the nation, France, form what we shall call an attribute (or characteristic) profile and can be shown graphically.

Figure 1

ATTRIBUTE PROFILE FOR FRANCE IN 1955*



*Plotted values are standard scores for France from five standardized characteristics on eighty-two nations. The plotted characteristics are:

ENERGY -- energy consumption per capita

INCOME -- national income

FREEOP -- freedom of group opposition

DENSTY -- population density

CATHLC -- ratio of catholics to the population

To illustrate grouping procedure, let us assume that we are interested in constructing a taxonomy from eighty-two nations using as our index of similarity the five characteristics of Figure 1. If, on a transparent material, we were to draw eighty-two separate profile graphs, one for each nation, we could then build a taxonomy simply by superimposing sheets of the transparent material. Suppose we took the graph for France shown in Figure 1 and superimposed another graph from one of the eighty-one remaining transparent sheets, continuing the process until we found the line which coincided most closely with that for France. Let us assume that the second graph is the attribute profile of West Germany. Taking the graphs for both France and West Germany, we superimpose a third profile from among the remaining eighty graphs continuing until we again find the one most similar to the two already grouped. As we continued this exercise we would observe that the lines of our superimposed profile group would spread increasingly over the sheet. We would conclude that as the number of nations in the group increases, the similarity of profile for that group would decrease.

Let us assume that we were interested in building a hierarchical taxonomy from the transparent sheets. We would first lay the eighty-two profiles side by side and then find the two profiles out of all combinations of two which were most similar and superimpose them. This would leave eighty-one profiles. We would again look for the two profiles most similar and again superimpose them. As we continued this exercise we would find that the visual criteria for similarity of profile would have to be relaxed. We would continue to relax the similarity criteria until all eighty-two profiles were superimposed. If we kept account of the order in which the profiles were grouped, we could draw a taxonomic map of our procedure similar to that of Figure 2.

For social scientists, however, the hierarchy of profile groupings may not be salient and we might try to build a cross-sectional taxonomy from the eighty-two profiles assuming static similarity. We could start by specifying the k number of groups we were interested in or we could specify the level of group coherence or similarity we were willing to accept and then see how many groups resulted. Either way, we would try the various permutations of superimposition until we found the best cross-section for our purposes.

The permutations at each step of our hypothetical grouping procedures would be so numerous for the eighty-two profile graphs that they would preclude actually attempting to construct taxonomies in this way. We must look to computer techniques for assistance.

III. APPLICATION

The transparent graph illustration presents grouping procedure as the matching of characteristic profiles. The available computer techniques do not match profiles, but instead reduce the profile statistics to single indices of similarity or distance between the cases to be grouped, and then match the indices. The interpretation of computer taxonomies has been difficult because the indices of similarity upon which the methods depend reflect, but do not reproduce the original profile characteristics.

In the transparent graph illustration we had attempted to group cases by their scores on five uncorrelated variables. The standard scales by which the attribute profile values were measured can be viewed as Cartesian coordinates of a five dimensional space. Each of the coordinate axis of this space is at right angles to all others, since the variables are uncorrelated. It then becomes possible to represent the

nation profiles as single points in the Cartesian coordinate system of the five variables. Each nation has a unique location in this space and this location is a unique Euclidean distance from every other point in the space. The similarity (congruence) of profiles is thus measured by the similarity of spatial location -- the distance between points. The question of profile similarity may thus be reduced to measuring the Euclidean distances within this space.

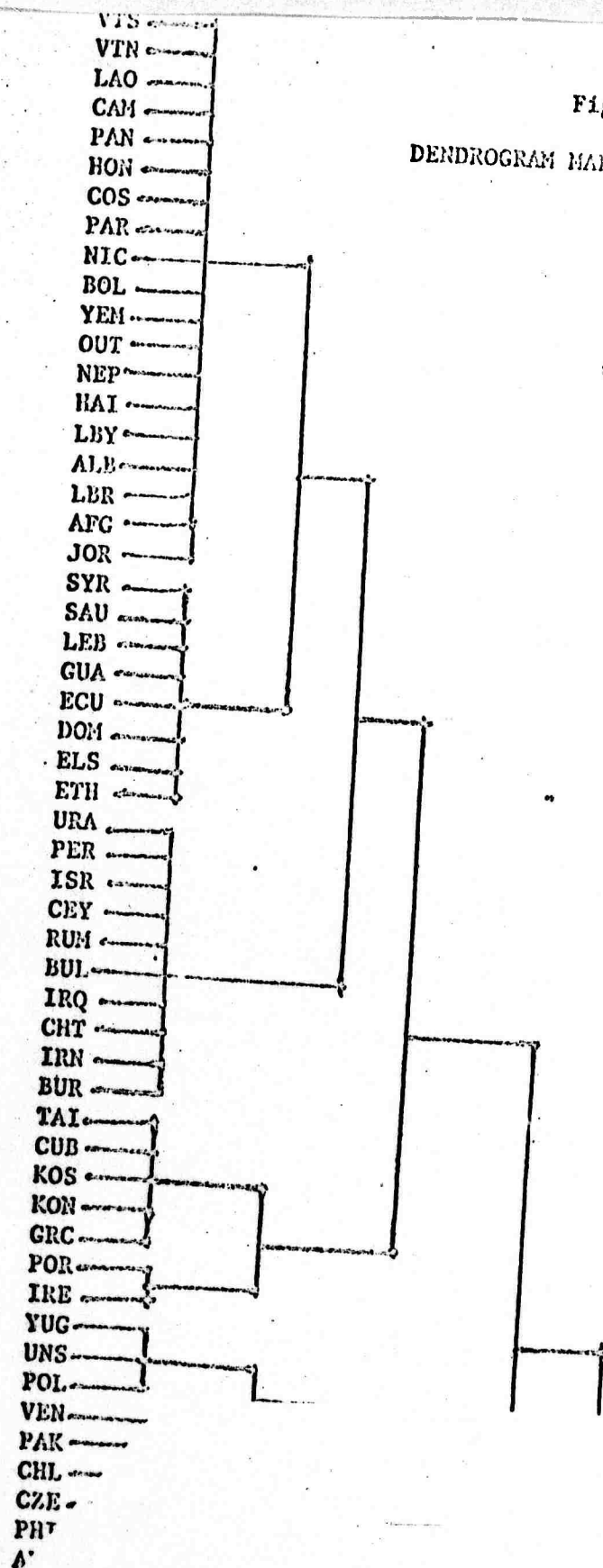
Since Euclidean distances between cases measure similarity, they can be used to develop taxonomies. Our concern at this point is not the development of a taxonomy, however, but with the underlying similarity the taxonomy represents.

A hierarchical taxonomic technique was employed to group the eighty-two nations on their Euclidean distances in the space of the five characteristics of Figure 1. A dendrogram displaying the resulting taxonomy is shown in Figure 2. The vertical lines represent the joining together of nations into a group. The distance between nations in the same group is increased as the procedure works from left to right in the dendrogram. At the extreme left no two nations are grouped and at the extreme right all nations form a single group.

As social scientists, we would probably not be interested in the entire dendrogram of Figure 2. Consequently, two interpretive problems arise. First, which groups should we extract from the dendrogram for presentation as our cross-section? Second, once we have selected our groups, how can we interpret the group similarities? To treat the latter problem we must return to the initial data -- to the profiles themselves.

Figure 2

DENDROGRAM MAPPING OF GROUPS*



*Grouping of 82 nations based on their distances in five uncorrelated characteristics. Grouping method was S. C. Johnson, Hierarchical Clustering Scheme, Diameter Method.

IV. GENERAL DESCRIPTION

Assume that we are interested in characterizing a group of k size given us by the taxonomic method. We calculate a group score for each of the original characteristics by adding together the member scores and dividing by k , the number of members in the group. This mean score we will call a group profile score for a characteristic. Since we normally find variation about the group mean score (group dissimilarity) we can calculate the group standard deviation as a measure of member deviation from the group mean. This variance will vary from characteristic to characteristic and will serve as a measure of group cohesion on a characteristic. If we assume the distribution of member scores around the mean profile score for a group to be normal, we can add a confidence interval to the group mean score. A one standard deviation confidence interval about the group profile score would encompass approximately two-thirds of the member scores for the group across the characteristics. A group profile with its confidence interval can be shown graphically. (Figure 3).

The horizontal midpoint of the plot in Figure 3 is the population mean value for the characteristics. We would expect mean profile scores for any group we extract from a population to tend toward the horizontal midpoint of the plot since this portion of the graph is the most dense portion of the variable scatter. If the group mean profile score on any one of the characteristics is far removed from the midpoint, then that characteristic distinguishes the group: group members are similar on the characteristic. The tight cluster of group scores on national income (INCOME) in Figure 3 is taken as the distinguishing characteristic of the plotted group.

Figure 3

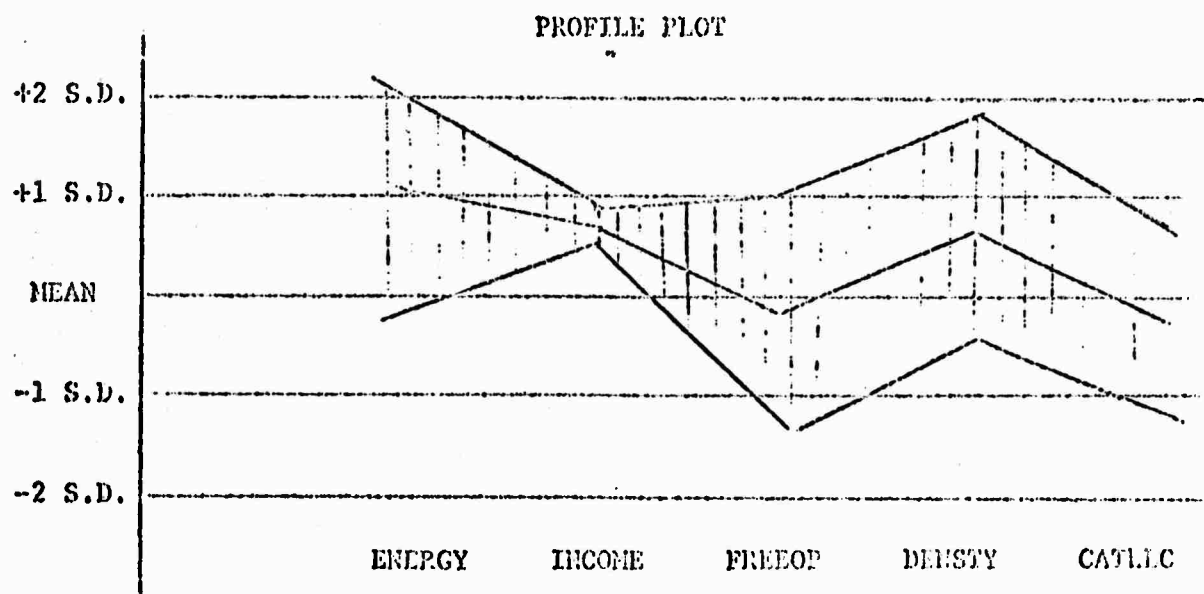
ATTRIBUTE PROFILE FOR GROUP FROM DENDROGRAM, FIG. 2*

Members of the group are:

France China West Germany Russia Great Britain

Profile Statistics for Five Selected Characteristics

	ENERGY	INCOME	FREEOP	DENSTY	CATHLC
AVERAGE SCORE	1.03	0.78	-0.15	0.64	-0.20
STANDARD DEVIATION	1.104	0.136	1.144	1.097	0.864
GROUP RANGE	3.4	0.3	2.3	2.7	2.3



*Group was derived from S.C. Johnson's Hierarchical Clustering Scheme, Diameter Method. Ref. S.C. Johnson, "Hierarchical Clustering Schemes," *Psychometrika*, vol. 32, No. 3, September 1967.

ENERGY = energy consumption per capita; INCOME = national income;
 FREEOP = freedom of group opposition; DENSTY = population density;
 CATHLC = proportion of catholics to the population.

V. D.O.N. COMPUTER PROGRAM PROFILE WRITEUP

1. DESCRIPTION

- A. This program calculates and plots group profiles.
- B. Input (data) up to 200 cases, 50 variables.
- C. Output
 - 1. input matrix
 - 2. means and standard deviations of input variables
 - 3. standardized data matrix (option)
 - 4. group profile statistics, up to 50 groups
 - 5. group plots with one standard deviation confidence interval (option)

2. ORDER OF CONTROL CARDS

A. Plot Symbol Card

- 1. col. 1-62 b.XX-9-8-7-6-5-4-3-2-1b0b+1+2+3+4+5+6+7+8+9101112131415161718

B. Main Control Card

- 1. col. 1-6 PROBLEM
- 2. col. 7-9 number of groups (maximum 50)
- 3. col. 10-12 number of cases (maximum 200)
- 4. col. 13-15 number of variables (maximum 50)
- 5. col. 18 1 standardize input data; blank do not standardize
- col. 21 1 plot group profiles; blank do not plot

C. Group Size Card (maximum size of group 50 cases)

- 1. col. 1-3 number of cases in first group
- 2. col. 4-6 number of cases in second group
- 3. col. 7-9 number of cases in third group
- 26. col. 76-78 number of cases in twenty-sixth group

(use a second card as needed to complete listing)

D. Observation Number Card (one set for each group)

- 1. col. 1-3 number of first case in the group
 - 2. col. 4-6 number of second case in the group
 - 3. col. 7-9 number of third case in the group
 - 26. col. 76-78 number of twenty-sixth case in the group
- (use a second card as needed to complete listing for the group)

E. Variable Format Card (use 2) both must be included even if second is blank.

F. Data Cards

G. Case Name Card

1. col. 1-6 name or number code of variable (as many cards as variables in the input data matrix. Blank cards must be inserted if variable names are not wanted)

I. Finish Card

1. col. 1-6 FINISH

3. MULTIPLE JOBS

- A. For a multiple job, repeat card sets (A) through (I) for each job. Card (I) signals the completion of all jobs and is placed at the end of the final job.

LEVEL 1, MOD 0

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THIS PROGRAM STANDARDIZES INPUT, CALCULATES MEAN, RANGE, STANDARD DEVIATION FOR GROUPS AND PLOTS THEIR PROFILE ACROSS THE VARIABLES. IT WILL ALSO CALCULATE THE PRODUCT-MOMENT CORRELATION COEFFICIENT BETWEEN GROUPS. IT MAY BE REDIMENSIONED AS NEEDED.

```

DIMENSION NSIZE(50), A(200,50), NOBS(50,200), AVG(50,50), STDEV(50,50)
*, RANGE(50,50), VENT(40), TMEAN(50), STAND(50), LINE(121), ZAVG(50,50), RPR
*(50,50), SYM(58)
DOUBLE PRECISION OBNAME(200), VNAME(50), NOB(50), FINISH, PROBLN
REAL LINE
DATA FINISH/'FINISH'/
FORMAT(A6,6I3)
FORMAT(26I3,2X)
FORMAT(20A4)
FORMAT(11X,121A1)
FORMAT(1X,12,1X,A6,1X,121A1)
FORMAT(11I1)
FORMAT(1X,13,1X,A6,11F11.3)
FORMAT(12X,'INPUT MATRIX A'//)
FORMAT(12X,'STANDARDIZED DATA MATRIX B'//)
FORMAT(12X,'MEAN AND STANDARD DEVIATION OF INPUT VARIABLES'//)
FORMAT(24X,'MEAN',12X,'STANDARD DEVIATION'//)
FORMAT(7X,13,2X,A6,2X,F10.3,14X,F14.3//)
FORMAT(2X,'GROUP PROFILE VALUES AND STATISTICS')
FORMAT(/12X,'MEMBERS OF GROUP',13)
FORMAT(/1X,'AVERAGE',4X,8(F10.4,3X))
FORMAT(/1X,'RANGE',6X,8(F10.4,3X))
FORMAT(/1X,'ST. DEV.',3X,8(F10.4,3X)/)
FORMAT(/17X,8(13,10X))
FORMAT(1P0,11(1X,12,1X,A6,1X))
FORMAT(/17X,11(13,8X))
FORMAT(1H1,'PROFILE PLOT OF GROUP',12,1X,'WITH ONE STANDARD DEVIAT
*ION CONFIDENCE INTERVAL'//)
FORMAT(/10X,121A1)
FORMAT(12X,'CHECK MEAN AND STANDARD DEVIATION OF STANDARDIZED MATR
*IX VARIABLES'//)
FORMAT(A6)
FORMAT(17X,11(A6,5X))
FORMAT(2X/2X)
FORMAT(17X,8(A6,7X))
FORMAT(63A1)
FORMAT(/17X,1X,'THE NUMBER OF STANDARD DEVIATIONS FROM THE PLOT LEFT
* MARGIN TO THE ORIGIN IS',12,1X,'THE NUMBER OF SPACES PER STANDARD
*/1X,'DEVIATION IS',13,1X,'THE ORIGIN IS LINE',13,1X,'THE RANGE OF
* DATA TO BE PLOTTED IS',F8.3,1X,'THE MINIMUM VALUE IS',F7.3,1X)
FORMAT(1H1,12X,'GROUP PRODUCT-MOMENT CORRELATION COEFFICIENTS'//)
FORMAT(11X,'NO.',1X,16I7//)
FORMAT(/6X,16,4X,16I7,3)

```

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	READ PLOTTING SYMBOLS	PROF0551
	READ(5,90) BLANK,DOT,X,AST,(SYM(J),J=1,58)	PROF0552
	READ IN NUMBER OF GROUPS, NUMBER OF OBSERVATIONS, NUMBER OF	PROF0553
	VARIABLES, AND OPTIONS TO STANDARDIZE INPUT, PLOT, AND CORRELATE	PROF0554
409	READ(5,10)PROBLEM,NGROUP,NCASE,NVAR,NSTAND,NPLOT,NCORR	PROF0570
	IF(PROBLEM.EQ.FINISH) GO TO 9999	PROF0571
	READ IN GROUP SIZES	PROF0580
	READ(5,11)(NSIZE(I),I=1,NGROUP)	PROF0581
	READ IN NUMBER OF EACH OBSERVATION IN EACH GROUP	PROF0600
	DO 1 I=1,NGROUP	PROF0610
	L=NSIZE(I)	PROF0620
	READ(5,11)(NOBS(I,J),J=1,L)	PROF0630
1	CONTINUE	PROF0640
	READ VARIABLE FORMAT	PROF0650
	READ(5,12) VFMT	PROF0651
	READ INPUT DATA	PROF0670
	DO 2 I=1,NCASE	PROF0680
	READ(5,VFMT)(A(I,J),J=1,NVAR)	PROF0690
2	CONTINUE	PROF0700
	READ(5,77)(OBNAME(J),J=1,NCASE)	PROF0710
	READ(5,77)(VNAME(J),J=1,NVAR)	PROF0720
	PRINT INPUT DATA	PROF0730
	NMIN=0	PROF0740
	I=(NCASE/51)+1	PROF0750
	JMOD=MOD(NVAR,11)	PROF0760
	IF(JMOD.EQ.0) J=(NVAR/11)	PROF0770
	IF(JMOD.NE.0) J=(NVAR/11)+1	PROF0780
	DO 402 II=1,I	PROF0790
	NMIN=NMIN+50	PROF0800
	DO 402 JJ=1,J	PROF0810
	KBEG=(JJ-1)*11+1	PROF0820
	NMIN=MIN0(NCASE,NMIN)	PROF0830
	KEND=MIN0((JJ*11),NVAR)	PROF0840
	WRITE(6,60)	PROF0850
	WRITE(6,61)	PROF0860
	IF(II.EQ.1) NBEG=1	PROF0870
	IF(II.EQ.2) NBEG=51	PROF0880
	IF(II.EQ.3) NBEG=101	PROF0890
	IF(II.EQ.4) NBEG=151	PROF0900
	WRITE(6,73)(IND,IND=KBEG,KEND)	PROF0910
	WRITE(6,76)(VNAME(IND),IND=KBEG,KEND)	PROF0920
	WRITE(6,80)	PROF0930
	DO 403 KK=NBEG,NMIN	PROF0940
	WRITE(6,50) KK,OBNAME(KK),(A(KK,IND),IND=KBEG,KEND)	PROF0950
03	CONTINUE	PROF0960
02	CONTINUE	PROF0970
	CALCULATE MEAN AND STANDARD DEVIATION OF INPUT DATA VARIABLES	PROF0980
	DO 7 J=1,NVAR	PROF0990

LEVEL 1, MOD 0

MAIN

DATE = 68117

10/24/13

	TOT=0.0	PROF1000
	TOTSQ=0.0	PROF1010
	DO 8 I=1,NCASE	PROF1020
	TOT=TOT+A(I,J)	PROF1030
	TOTSQ=TOTSQ+(A(I,J)**2)	PROF1040
8	CONTINUE	PROF1050
	TMEAN(J)=TOT/LOAT(NCASE)	PROF1060
	IF((TOTSQ/LOAT(NCASE))-(TMEAN(J)**2).LE.0.0) GO TO 87	PROF1070
	STAND(J)=SQRT((TOTSQ/LOAT(NCASE))-(TMEAN(J)**2))	PROF1080
87	CONTINUE	PROF1090
7	CONTINUE	PROF1100
	PRINT MEAN AND STANDARD DEVIATION OF INPUT DATA VARIABLES	PROF1110
	WRITE(6,60)	PROF1120
	WRITE(6,63)	PROF1130
	WRITE(6,64)	PROF1140
	WRITE(6,65)(J,VNAME(J),TMEAN(J),STAND(J),J=1,NVAR)	PROF1150
	TEST FOR STANDARDIZING INPUT DATA MATRIX OPTION	PROF1160
	IF(NSIAND.NE.1) GO TO 99	PROF1180
	STANDARDIZE INPUT VARIABLES	PROF1190
	DO 15 J=1,NVAR	PROF1200
	DO 9 I=1,NCASE	PROF1210
9	A(I,J)=(A(I,J)-TMEAN(J))/STAND(J)	PROF1220
45	CONTINUE	PROF1230
	PRINT STANDARDIZED DATA MATRIX B	PROF1240
	NMIN = 0	PROF1250
	I=(NCASE/5)+1	PROF1260
	JMOD=MOD(NVAR,11)	PROF1270
	IF(JMOD.EQ.0) J=(NVAR/11)	PROF1280
	IF(JMOD.NE.0) J=(NVAR/11)+1	PROF1290
	DO 400 I1=1,I	PROF1300
	NMIN=NRIN+50	PROF1310
	DO 400 JJ = 1,J	PROF1320
	KBEG=(JJ-1)*11+1	PROF1330
	NMIN=MINO(NCASE,NMIN)	PROF1340
	KEND=MINO((JJ*11),NVAR)	PROF1350
	WRITE(6,60)	PROF1360
	WRITE(6,62)	PROF1370
	IF(I1.EQ.1) NBEG=1	PROF1380
	IF(I1.EQ.2) NBEG=51	PROF1390
	IF(I1.EQ.3) NBEG=101	PROF1400
	IF(I1.EQ.4) NBEG=151	PROF1410
	WRITE(6,73)(IND,IND=KBEG,KEND)	PROF1420
	WRITE(6,78)(VNAME(IND),IND=KBEG,KEND)	PROF1430
	WRITE(6,80)	PROF1440
	DO 401 KK=NBEG,NMIN	PROF1450
	WRITE(6,50)KK,UBNAME(KK),(A(KK,IND),IND=KBEG,KEND)	PROF1460
401	CONTINUE	PROF1470

LEVEL 1, MOD 0	MAIN	DATE = 68117	10/24/13
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400	CONTINUE	PROF1400
99	CONTINUE	PROF1400
C		PROF1500
C	CALCULATE GROUP MEAN, STANDARD DEVIATION, AND RANGE	PROF1510
	BMAX=-9999.0	PROF1520
	BMIN=9999.0	PROF1530
	DO 3 I = 1,NGROUP	PROF1540
	DO 4 J = 1,NVAR	PROF1550
	TOT=0.0	PROF1560
	TOTSQ=0.0	PROF1570
	AMAX=-9999.0	PROF1580
	AMIN=9999.0	PROF1590
	NSZ=NSIZE(1)	PROF1600
	DO 5 K=1,NSZ	PROF1610
	L = NOBS(1,K)	PROF1620
	TOT=TOT+A(L,J)	PROF1630
	TOTSQ=TOTSQ+(A(L,J)**2)	PROF1640
	IF(AMAX.LT.A(L,J)) AMAX = A(L,J)	PROF1650
	IF(AMIN.GT.A(L,J)) AMIN = A(L,J)	PROF1660
5	CONTINUE	PROF1670
	AVG(1,J) = TOT/FLOAT(NSIZE(1))	PROF1680
	STDEV(1,J)=0.0	PROF1690
	IF((TOTSQ/FLOAT(NSIZE(1)))-(AVG(1,J)**2).LE.0.0) GO TO 85	PROF1700
	STDEV(1,J)=SQRT((TOTSQ/FLOAT(NSIZE(1)))-(AVG(1,J)**2))	PROF1710
85	CONTINUE	PROF1720
	IF(BMAX.LT.(AVG(1,J)+STDEV(1,J))) BMAX=AVG(1,J)+STDEV(1,J)	PROF1730
	IF(BMIN.GT.(AVG(1,J)-STDEV(1,J))) BMIN=AVG(1,J)-STDEV(1,J)	PROF1740
4	RANGE(1,J) = AMAX-AMIN	PROF1750
3	CONTINUE	PROF1760
	RANG=BMAX+ABS(BMIN)	PROF1770
C	PRINT GROUP MEAN, STANDARD DEVIATION, AND RANGE	PROF1780
	WRITE(6,60)	PROF1790
	WRITE(6,66)	PROF1800
	DO 405 I=1,NGROUP	PROF1810
	WRITE(6,67) I	PROF1820
	NSZ = NSIZE(I)	PROF1830
	DO 6 J=1,NSZ	PROF1840
	N=NOBS(I,J)	PROF1850
6	NOB(J)=OBNAME(N)	PROF1860
	WRITE(6,72)(NOBS(1,J),NOB(J),J=1,NSZ)	PROF1870
	JMOD=MOD(NVAR,8)	PROF1880
	IF(JMOD.EQ.0) J = NVAR/8	PROF1890
	IF(JMOD.NE.0) J = (NVAR/8)+1	PROF1900
	DO 404 JJ = 1,J	PROF1910
	KBEG=((JJ-1)*8)+1	PROF1920
	KEND=MIN(1+JJ*8,NVAR)	PROF1930
	WRITE(6,71)(N,N=KBEG,KEND)	PROF1940
	WRITE(6,81)(VNAME(I),N=KBEG,KEND)	PROF1950

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WRITE(6,68) (AVG(1,N),N=KBEG,KEND)
 WRITE(6,70) (STDEV(1,N),N=KBEG,KEND)
 WRITE(6,69) (RANGE(1,N),N=KBEG,KEND)

PROF 187
 PROF 188
 PROF 189

404 CONTINUE

PROF 190

K=0

PROF 191

DO 406 11=1,12

PROF 201

K=K+4

PROF 201

IF(1.NE.K) GO TO 406

PROF 201

WRITE(6,60)

PROF 202

406 CONTINUE

PROF 203

405 CONTINUE

PROF 204

IF(NPLOT.NE.1) GO TO 977

PROF 205

PLOT GROUP MEAN ACROSS VARIABLES WITH ONE S.DEV. CONFIDENCE

PROF 206

INTERVAL ROUNDING THE PLOT LOCATIONS

PROF 207

KSP=100.0/RANG

PROF 208

LA=ABS(BMIN)+1.0

PROF 209

KOR=(KSP*LA)+1

PROF 210

WRITE(6,106) LA,KSP,KOR,RANG,BMIN

PROF 211

DO 999 1=1,NGROUP

PROF 212

DO 105 J=1,121

PROF 213

105 LINE(J)=BLANK

PROF 214

N=20-(2*LA)

PROF 215

DO 51 11=1,121,KSP

PROF 216

M=N-1

PROF 217

J=11+1

PROF 218

LINE(11)=SYM(M)

PROF 219

LINE(J)=SYM(N)

PROF 220

51 N=N+2

PROF 221

IF(LA.GT.9) LINE(KOR)=SYM(20)

PROF 222

WRITE(6,74) 1

PROF 223

NSZ=NSIZE(1)

PROF 224

DO 52 J=1,NSZ

PROF 225

N=NOBS(1,J)

PROF 226

52 NOB(J)=OBNAME(N)

PROF 227

WRITE(6,72) (NOBS(1,J),NOB(J),J=1,NSZ)

PROF 228

WRITE(6,75) LINE

PROF 229

DO 101 J=1,121

PROF 230

101 LINE(J)=D01

PROF 231

WRITE(6,31) LINE

PROF 232

DO 103 J=1,121

PROF 233

103 LINE(J)=BLANK

PROF 234

DO 190 NN=1,121,KSP

PROF 235

190 LINE(KN)=D01

PROF 236

DO 30 K=1,NVAR

PROF 237

K=((AVG(1,K)*FLOAT(KSP))+0.5)+FLOAT(KOR)

PROF 238

KK=((((AVG(1,K)+STDEV(1,K))*FLOAT(KSP))+0.5)+FLOAT(KOR)

PROF 239

KK=((((AVG(1,K)-STDEV(1,K))*FLOAT(KSP))+0.5)+FLOAT(KOR)

PROF 240

NOT REPRODUCIBLE

LEVEL 1, MOD 0

MAIN

DATE = 68117

10/24/15

	LINE(M)=X	PROF 251
	IF(MN.GT.120) GO TO 179	PROF 251
	LINE(MN)=AS1	PROF 251
179	CONTINUE	PROF 251
	IF(MN.LT.0) GO TO 79	PROF 251
	LINE(MN)=AS1	PROF 251
79	CONTINUE	PROF 251
	WRITE(6,32)K,VNAME(K), LINE	PROF 251
	DO 104 J=1,121	PROF 251
104	LINE(J)=BLANK	PROF 251
	DO 191 NN=1,121,KSP	PROF 251
191	LINE(NN)=DOT	PROF 251
	DO 988 JJ=1,4	PROF 251
	WRITE(6,31) LINE	PROF 251
988	CONTINUE	PROF 251
30	CONTINUE	PROF 251
999	CONTINUE	PROF 251
977	CONTINUE	PROF 251
	GO TO 409	PROF 251
9999	CONTINUE	PROF 251
	STOP	PROF 251
	END	PROF 251

VI. SAMPLE PROBLEM

To characterize a cross-sectional taxonomy from the dendrogram of Figure 2, we selected a cross-section of eight groups and plotted their profiles with the profile program. The program features an automatic scaling device which scales the plots to the range of the statistics to be plotted in each job. The United States, the sole member of the eighth group, proved to be an extreme outlier on national income and energy consumption. To demonstrate the variable scale feature, groups 1 and 2 were plotted by themselves, without the necessity to scale the plots for the United States. The Plots for Groups 3 and 6 are products of the computer run when the United States was plotted. The computer output for these plots and statistics for the first three groups are shown on the following pages.

To allow for a large number of variables, plots are made vertically down the page, beginning with variable 1 at the top. The group mean score is denoted by an "X" and "*" denotes the one standard deviation confidence interval about the mean score. If a single asterisk appears, as in the group 1 plot for INCOME, member standard deviation is negligible and the confidence interval converges on a single point. If the asterisk is very close to the other asterisk, there may be no "X" between them. It is assumed that the mean score would fall between them in this case. The user may wish to shade the confidence intervals in the output as an aid in interpreting the profile variation across characteristics for the group.

GROUP PROFILE VALUES AND STATISTICS

MEMBERS OF GROUP 1

26 ETH	25 ILS	22 DON	23 ECU	32 GLA	47 LEB	65 SAU	69 SYR	44 JOR	1 APR
2 ALB	49 LBY	33 HAI	51 NEP	56 OUT	78 YEM	7 BUL	54 NIC	59 PAR	18 CCS
38 PAR	11 CAM	30 LAO	81 VTN	82 VTS	48 LBR	34 HON			

1	2	3	4	5
ENERGY	INCOME	FREEDP	DENSITY	CATHLC

AVERAGE -0.6257 -0.2423 -0.1250 -0.4162 0.0019

ST. DEV. 0.0777 0.0047 1.0218 0.5151 1.0753

RANGE 0.5160 0.0161 2.3359 1.8518 2.3565

MEMBERS OF GROUP 2

10 BUR	30 IRN	16 CHI	39 IRQ	9 BUL	64 ROM	15 CEY	41 ISR	60 PER	76 URA
--------	--------	--------	--------	-------	--------	--------	--------	--------	--------

1	2	3	4	5
ENERGY	INCOME	FREEDP	DENSITY	CATHLC

AVERAGE -0.6730 -0.2239 -0.1510 0.0324 -0.4238

ST. DEV. 0.2520 0.0030 1.0182 0.5714 0.8283

RANGE 0.7423 0.0082 2.3359 3.2000 2.1423

MEMBERS OF GROUP 3

40 IND	53 PER	81 ORG	45 KON	46 KOS	19 CUB	70 TAI
--------	--------	--------	--------	--------	--------	--------

1	2	3	4	5
ENERGY	INCOME	FREEDP	DENSITY	CATHLC

AVERAGE -0.5121 -0.2041 -0.2177 1.0009 0.0277

ST. DEV. 0.0000 0.0000 0.9749 1.0000 1.1044

RANGE 0.0000 0.0000 2.3359 2.0000 2.3062

NOT REPRODUCIBLE

PROFILE PLOT OF GROUP 1 WITH ONE STANDARD DEVIATION CONFIDENCE INTERVAL

	25 ELS	22 DUM	23 ECU	32 GUA	47 LEB	65 SAU	09 SYR	44 JOR	1 AFG
2 ALB	49 LBY	53 HAI	51 NEP	56 OUT	78 YEM	7 BOL	54 NIC	59 PAR	18 COS
58 PAN	11 CAN	60 LAO	61 VIN	82 VTS	48 LBR	34 HON			

-2

-1

0

+1

1 ENERGY

2 INCOME

3 FREED

4 DENSITY

5 CATHOL

NOT REPRODUCIBLE

PROFILE PLOT OF GROUP 2 WITH ONE STANDARD DEVIATION CONFIDENCE INTERVAL

10 BUK	38 IRG	16 CHI	39 IRQ	9 BUL	64 RUM	13 CEY	41 ISR	60 PER	76 URA
--------	--------	--------	--------	-------	--------	--------	--------	--------	--------

-2

-1

0

+1

1 ENERGY

2 INCOME

3 FREEDOM

4 DENSITY

5 CATHOLIC

NOT REPRODUCIBLE

PROFILE PLOT OF GROUP 3 WITH ONE STANDARD DEVIATION CONFIDENCE INTERVAL

40 IRE 03 PGR 31 GRC 45 KON 46 KOS 49 CUB 70 TAI

-2 -1 0 +1 +2 +3 +4 +5 +6 +7

1 ENERGY

.....

2 INCOME

.....

3 FREEDOM

.....

4 DENSITY

.....

5 CATHOLIC

.....

NOT REPRODUCIBLE

PROFILE PLOT OF GROUP 6 WITH ONE STANDARD DEVIATION CONFIDENCE INTERVAL

12 JAN 36 IND 42 ITA 43 JAP

-2 -1 0 +1 +2 +3 +4 +5 +6 +7

1 ENERGY

.....#.....X.....#.....

2 INCOME

.....**.....

3 FREED

.....*.....

4 DENSITY

.....*.....X.....*

5 CATHOL

.....*.....X.....*

NOT REPRODUCIBLE